

# Constrained Optimization



**Prof (Dr.) Rajib Kumar Bhattacharjya**

Department of Civil Engineering  
INDIAN INSTITUTE OF TECHNOLOGY, GUWAHATI  
GUWAHATI-781039, ASSAM, INDIA

# Constrained Optimization

Multivariable problem with inequality constraints

Minimize  $f(X)$       Where  $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to  $g_j(X) \leq 0$        $j = 1, 2, 3, \dots, m$

We can write  $g_j(X) + y_j^2 = 0$

Thus the problem can be written as

Minimize  $f(X)$

Subject to  $G_j(X, Y) = g_j(X) + y_j^2 = 0$        $j = 1, 2, 3, \dots, m$

Where  $Y = [y_1, y_2, y_3, \dots, y_m]^T$

# Constrained Optimization

Multivariable problem with inequality constraints

Minimize  $f(X)$       Where  $X = [x_1, x_2, x_3, \dots, x_n]^T$

Subject to  $G_j(X, Y) = g_j(X) + y_j^2 = 0$        $j = 1, 2, 3, \dots, m$

The Lagrange function can be written as

$$L(X, Y, \lambda) = f(X) + \sum_{j=1}^m \lambda_j G_j(X, Y)$$

The necessary conditions of optimality can be written as

$$\frac{\partial L(X, Y, \lambda)}{\partial x_i} = \frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$\frac{\partial L(X, Y, \lambda)}{\partial \lambda_j} = G_j(X, Y) = g_j(X) + y_j^2 = 0 \quad j = 1, 2, 3, \dots, m$$

$$\frac{\partial L(X, Y, \lambda)}{\partial y_j} = 2\lambda_j y_j = 0 \quad j = 1, 2, 3, \dots, m$$

# Constrained Optimization

Multivariable problem with inequality constraints

From equation  $\frac{\partial L(X,Y,\lambda)}{\partial y_j} = 2\lambda_j y_j = 0$

Either  $\lambda_j = 0$  Or,  $y_j = 0$

If  $\lambda_j = 0$ , the constraint is not active, hence can be ignored

If  $y_j = 0$ , the constraint is active, hence have to consider

Now, consider all the active constraints,

Say set  $J_1$  is the active constraints

And set  $J_2$  is the inactive constraints

The optimality condition can be written as

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j \in J_1} \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$g_j(X) = 0 \quad j \in J_1$$

$$g_j(X) + y_j^2 = 0 \quad j \in J_2$$

# Constrained Optimization

## Multivariable problem with inequality constraints

$$-\frac{\partial f}{\partial x_i} = \lambda_1 \frac{\partial g_1}{\partial x_i} + \lambda_2 \frac{\partial g_2}{\partial x_i} + \lambda_3 \frac{\partial g_3}{\partial x_i} + \dots + \lambda_p \frac{\partial g_p}{\partial x_i} \quad i = 1, 2, 3, \dots, n$$

$$-\nabla f = \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2 + \lambda_3 \nabla g_3 + \dots + \lambda_p \nabla g_p$$

This indicates that negative of the gradient of the objective function can be expressed as a linear combination of the gradient of the active constraints at optimal point.

$$\nabla f = \begin{Bmatrix} \partial f / \partial x_1 \\ \partial f / \partial x_2 \\ \vdots \\ \partial f / \partial x_n \end{Bmatrix}$$

$$\nabla g_j = \begin{Bmatrix} \partial g_j / \partial x_1 \\ \partial g_j / \partial x_2 \\ \vdots \\ \partial g_j / \partial x_n \end{Bmatrix}$$

$$-\nabla f = \lambda_1 \nabla g_1 + \lambda_2 \nabla g_2$$

Let  $S$  be a feasible direction, then we can write

$$-S^T \nabla f = \lambda_1 S^T \nabla g_1 + \lambda_2 S^T \nabla g_2$$

Since  $S$  is a feasible direction

$$S^T \nabla g_1 < 0 \quad \text{and} \quad S^T \nabla g_2 < 0$$

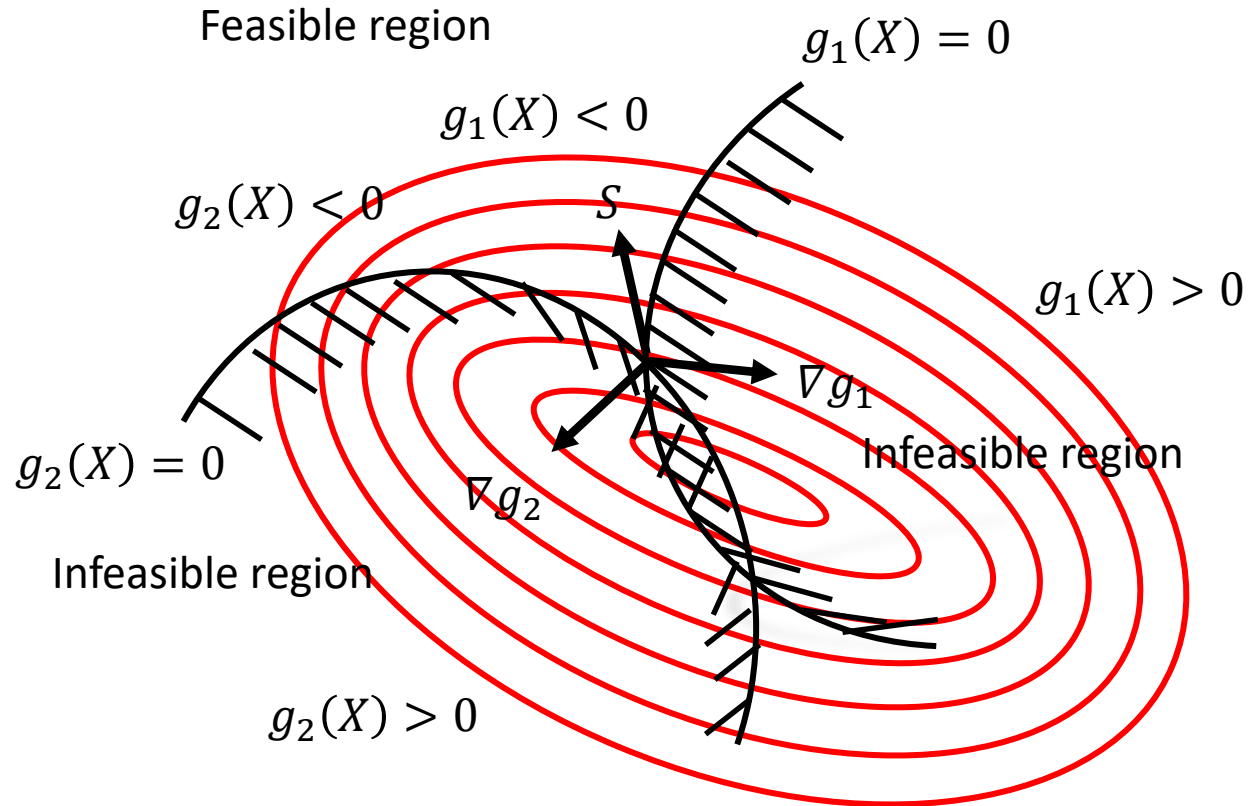
If  $\lambda_1, \lambda_2 > 0$

Then the term  $S^T \nabla f$  is +ve

This indicates that  $S$  is a direction of increasing function value

Thus we can conclude that if  $\lambda_1, \lambda_2 > 0$ , we will not get any better solution than the current solution

# Constrained Optimization



$$-S^T \nabla f = \lambda_1 S^T \nabla g_1 + \lambda_2 S^T \nabla g_2$$

Since  $S$  is a feasible direction  
 $S^T \nabla g_1 < 0$  and  $S^T \nabla g_2 < 0$

If  $\lambda_1, \lambda_2 > 0$   
Then the term  $S^T \nabla f$  is +ve

This indicates that  $S$  is a direction  
of increasing function value

Thus we can conclude that if  
 $\lambda_1, \lambda_2 > 0$ , we will not get any  
better solution than the current  
solution

# Constrained Optimization

Multivariable problem with inequality constraints

The necessary conditions to be satisfied at constrained minimum points  $X^*$  are

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j \in J_1} \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$
$$\lambda_j \geq 0 \quad j \in J_1$$

These conditions are called **Kuhn-Tucker conditions**, the necessary conditions to be satisfied at a relative minimum of  $f(X)$ .

These conditions are in general not sufficient to ensure a relative minimum, However, in case of a convex problem, these conditions are the necessary and sufficient conditions for global minimum.

# Constrained Optimization

## Multivariable problem with inequality constraints

If the set of active constraints are not known, the Kuhn-Tucker conditions can be stated as

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$\left. \begin{array}{l} \lambda_j g_j = 0 \\ g_j \leq 0 \\ \lambda_j \geq 0 \end{array} \right\} \quad j = 1, 2, 3, \dots, m$$



# Constrained Optimization

Multivariable problem with equality and inequality constraints

For the problem

$$\begin{array}{lll} \text{Minimize} & f(X) & \text{Where } X = [x_1, x_2, x_3, \dots, x_n]^T \\ \text{Subject to} & g_j(X) \leq 0 & j = 1, 2, 3, \dots, m \\ & h_k(X) = 0 & k = 1, 2, 3, \dots, p \end{array}$$

The Kuhn-Tucker conditions can be written as

$$\frac{\partial f(X)}{\partial x_i} + \sum_{j=1}^m \lambda_j \frac{\partial g_j(X)}{\partial x_i} + \sum_{k=1}^p \beta_k \frac{\partial h_k(X)}{\partial x_i} = 0 \quad i = 1, 2, 3, \dots, n$$

$$\lambda_j g_j = 0 \quad j = 1, 2, 3, \dots, m$$

$$g_j \leq 0 \quad j = 1, 2, 3, \dots, m$$

$$h_k = 0 \quad k = 1, 2, 3, \dots, p$$

$$\lambda_j \geq 0 \quad j = 1, 2, 3, \dots, m$$

# Constrained Optimization

$$\textit{Minimize } f(X) = x_1^2 + 2x_2^2 + 3x_3^2$$

$$\textit{Subject to } g_1(X) = x_1 - x_2 - 2x_3 \leq 12$$

$$g_2(X) = x_1 + 2x_2 - 3x_3 \leq 8$$